

Multiparent populations & R/qtl2

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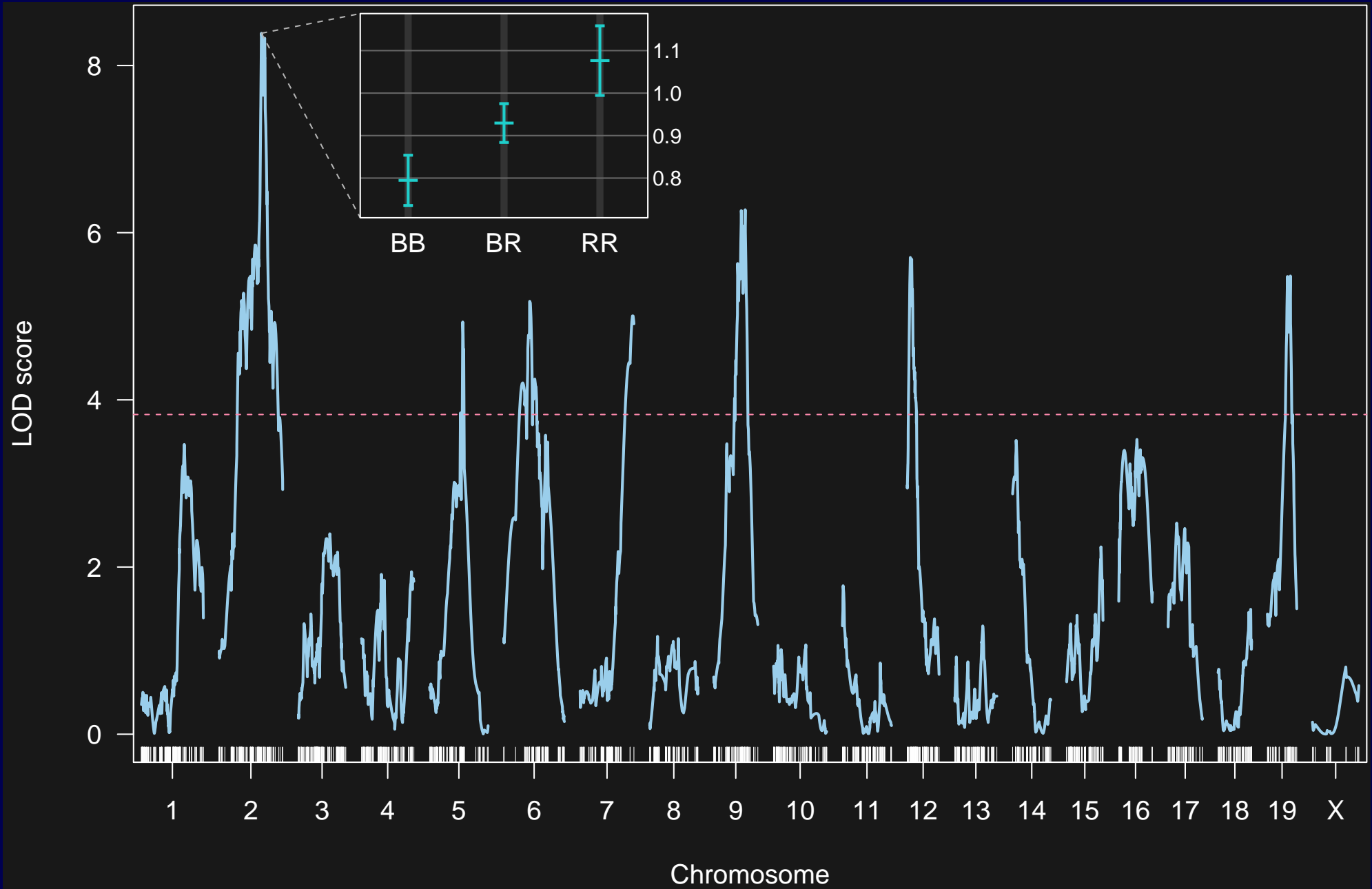
kbroman.org/qtl2

kbroman.org

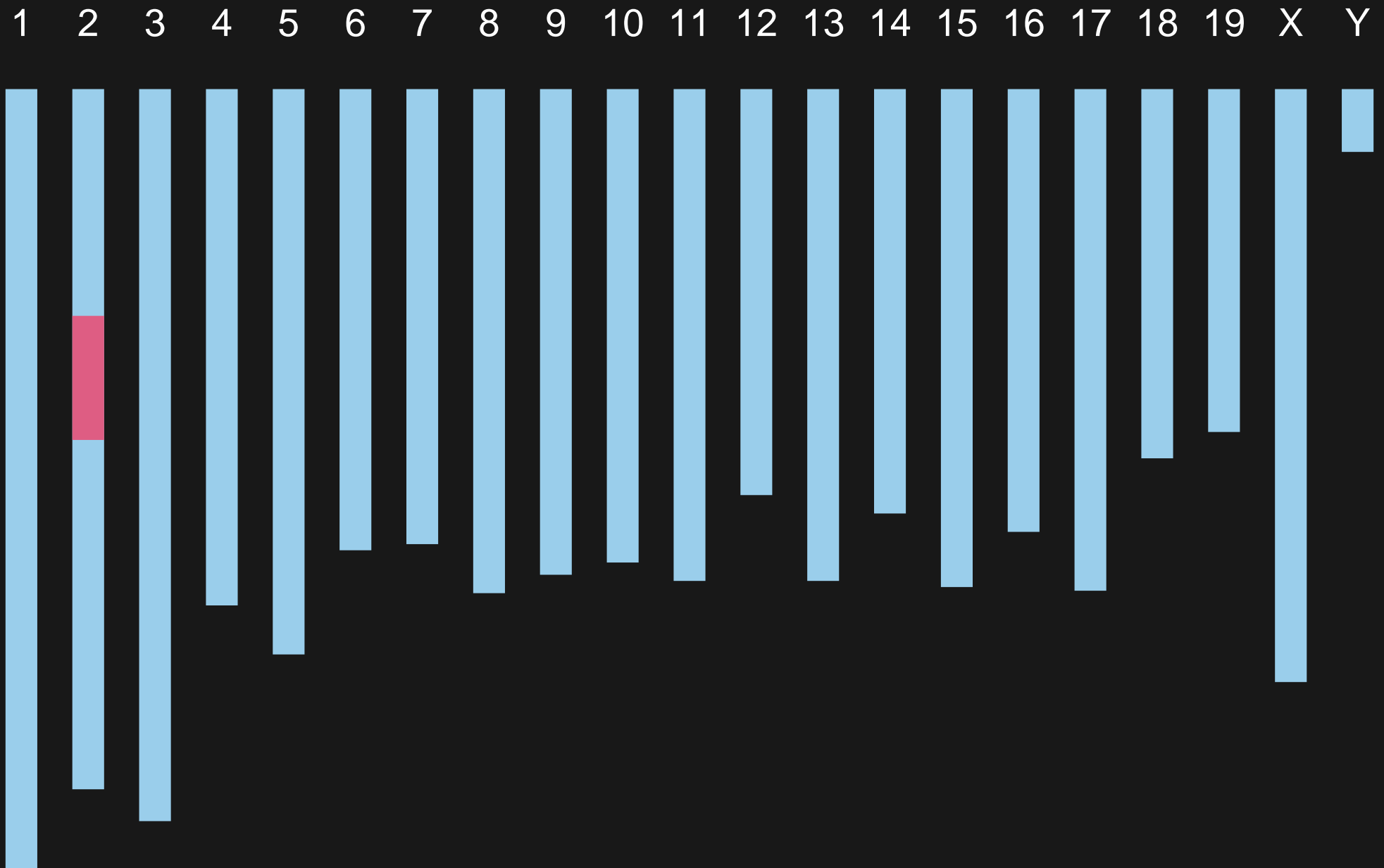
github.com/kbroman

[@kwbroman](https://twitter.com/kwbroman)

QTL mapping



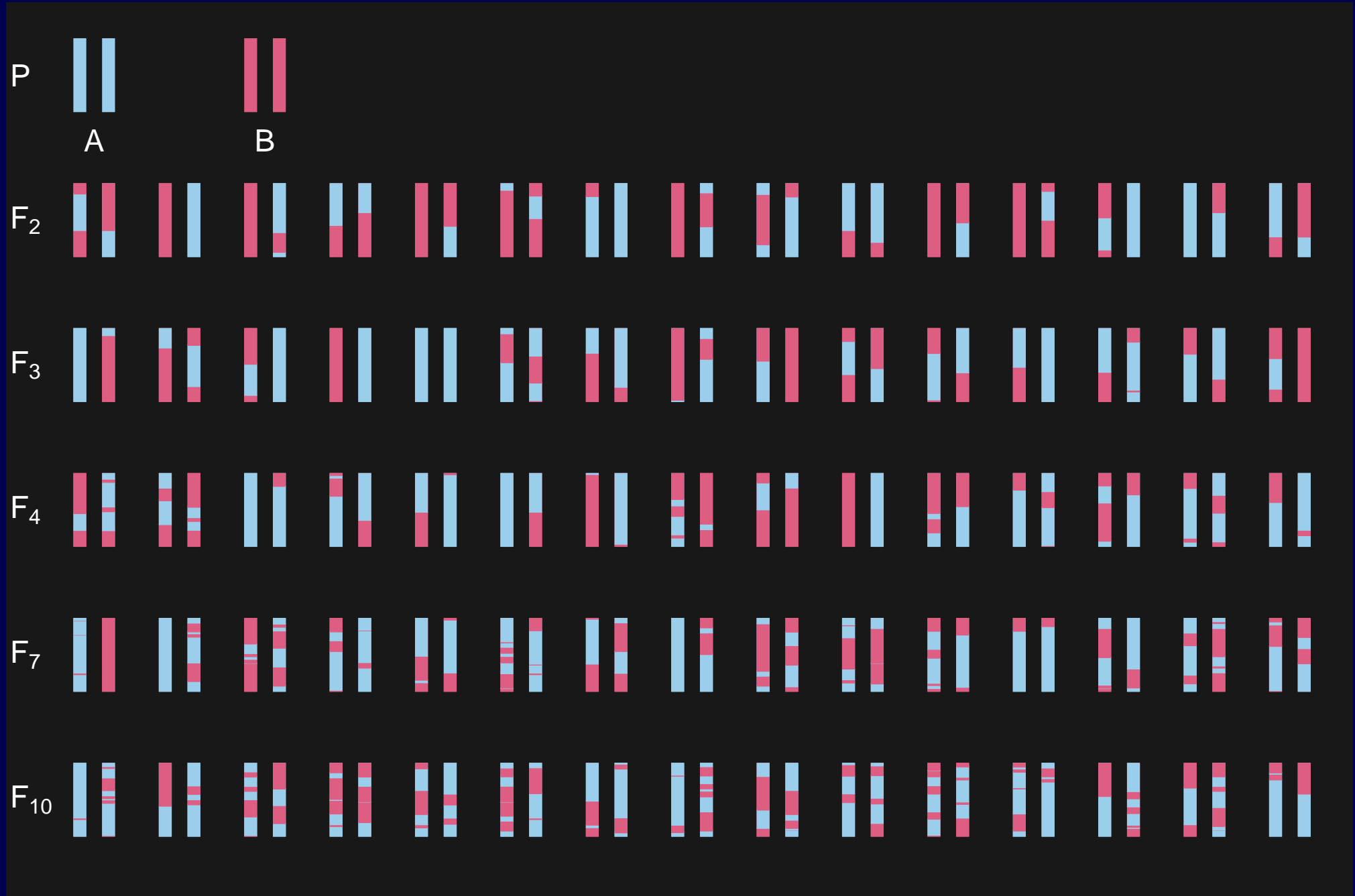
Congenetic line / NIL



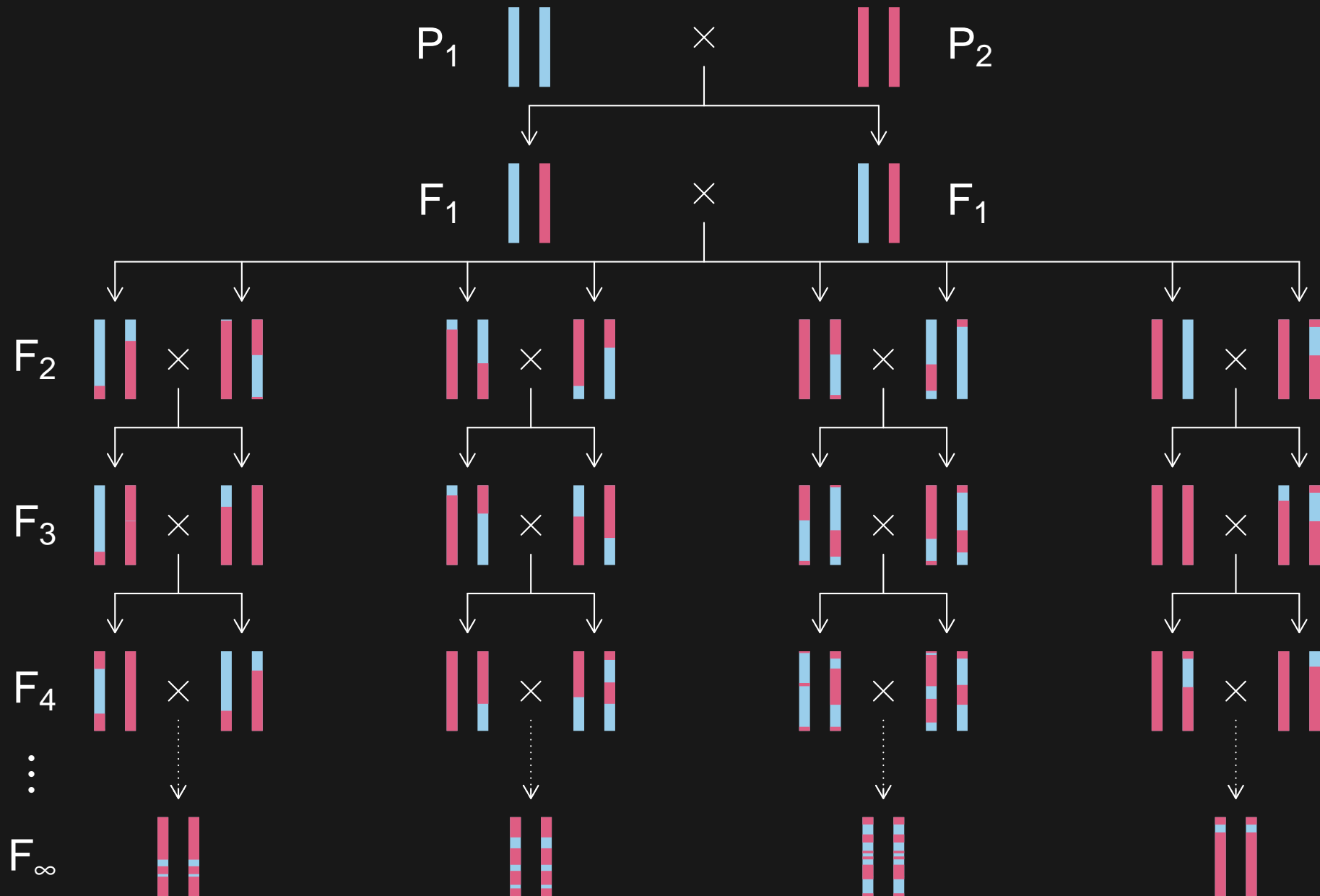
Improving precision

- more recombinations
- more individuals
- more precise phenotypes
- lower-level phenotypes
transcripts, proteins, metabolites

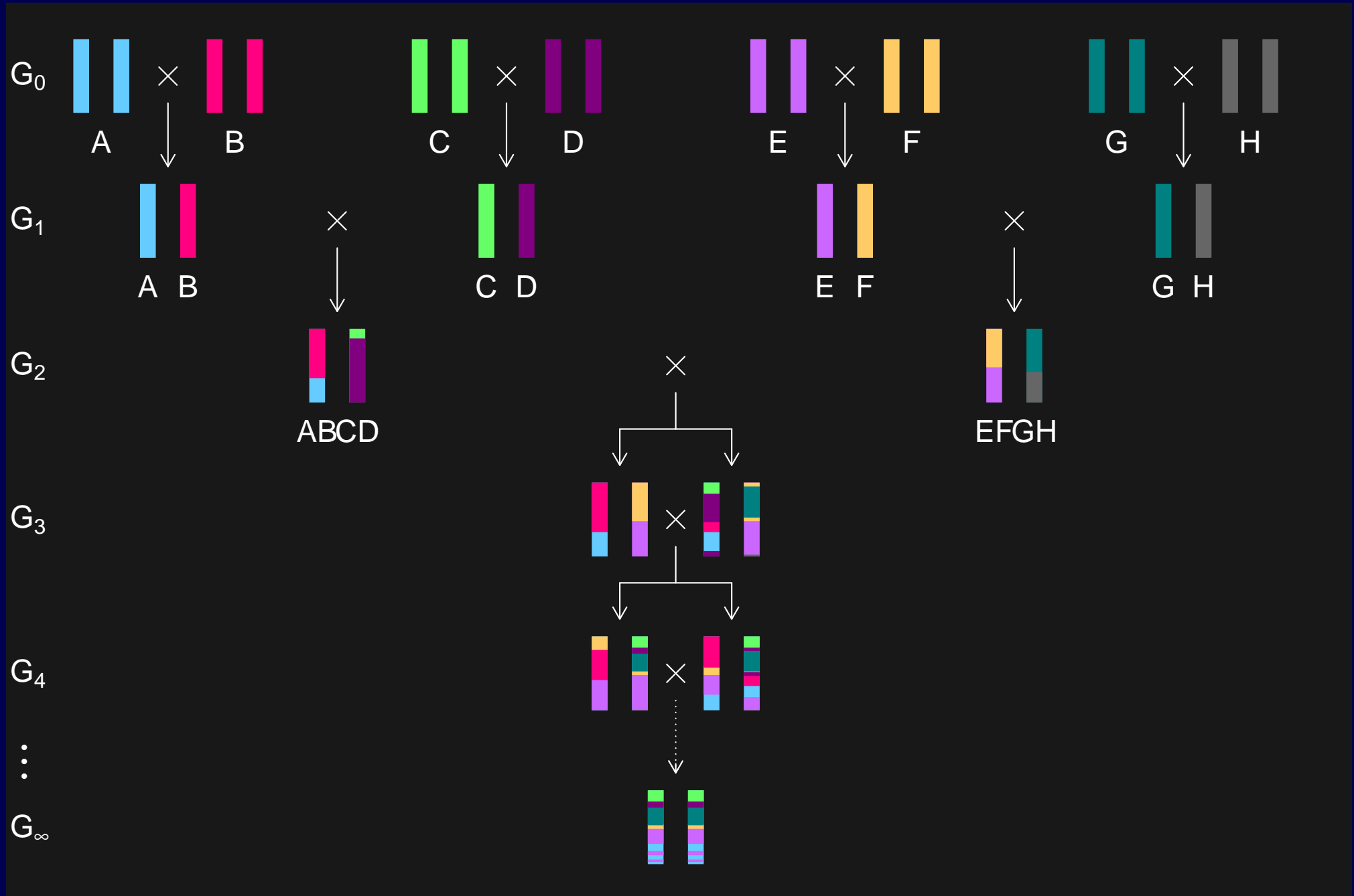
Advanced intercross lines



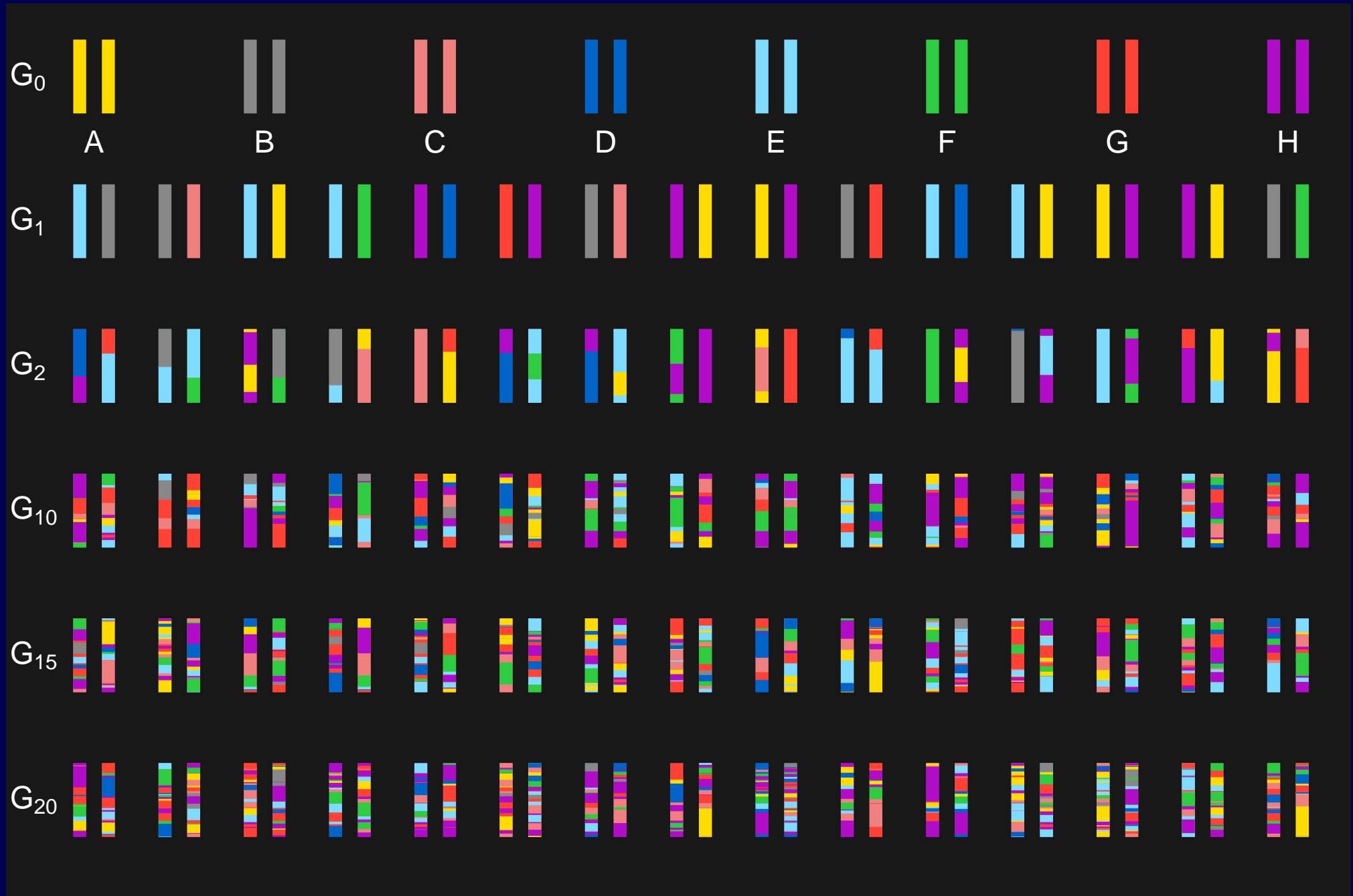
Recombinant inbred lines



Collaborative Cross/MAGIC



Heterogeneous stock



Why R/qtl2?

- High-dimensional data
genotypes and phenotypes
- More diverse crosses
especially multi-parent populations
- Linear mixed models
especially in DO/HS/AIL

R/qlt → R/qlt2

- See kbroman.org/qlt2/assets/vignettes/rqlt_diff.html
- New data file formats
- New data structures
- New function names

`read.cross()` → `read_cross2()`

`calc.genoprob()` → `calc_genoprob()`

`scanone()` → `scan1()`

- Different treatment of intermediate calculations
- Use of individual IDs for aligning data
- Order of args when subsetting cross objects

`cross[chr,ind]` → `cross2[ind,chr]`

→ R

- `convert2cross2()`
- `summary()`, `n_ind()`, `n_mar()`, ...
- `insert_pseudomarkers()`
- `calc_genoprob()`
- `scan1()`
- `find_peaks()`

Linear mixed models

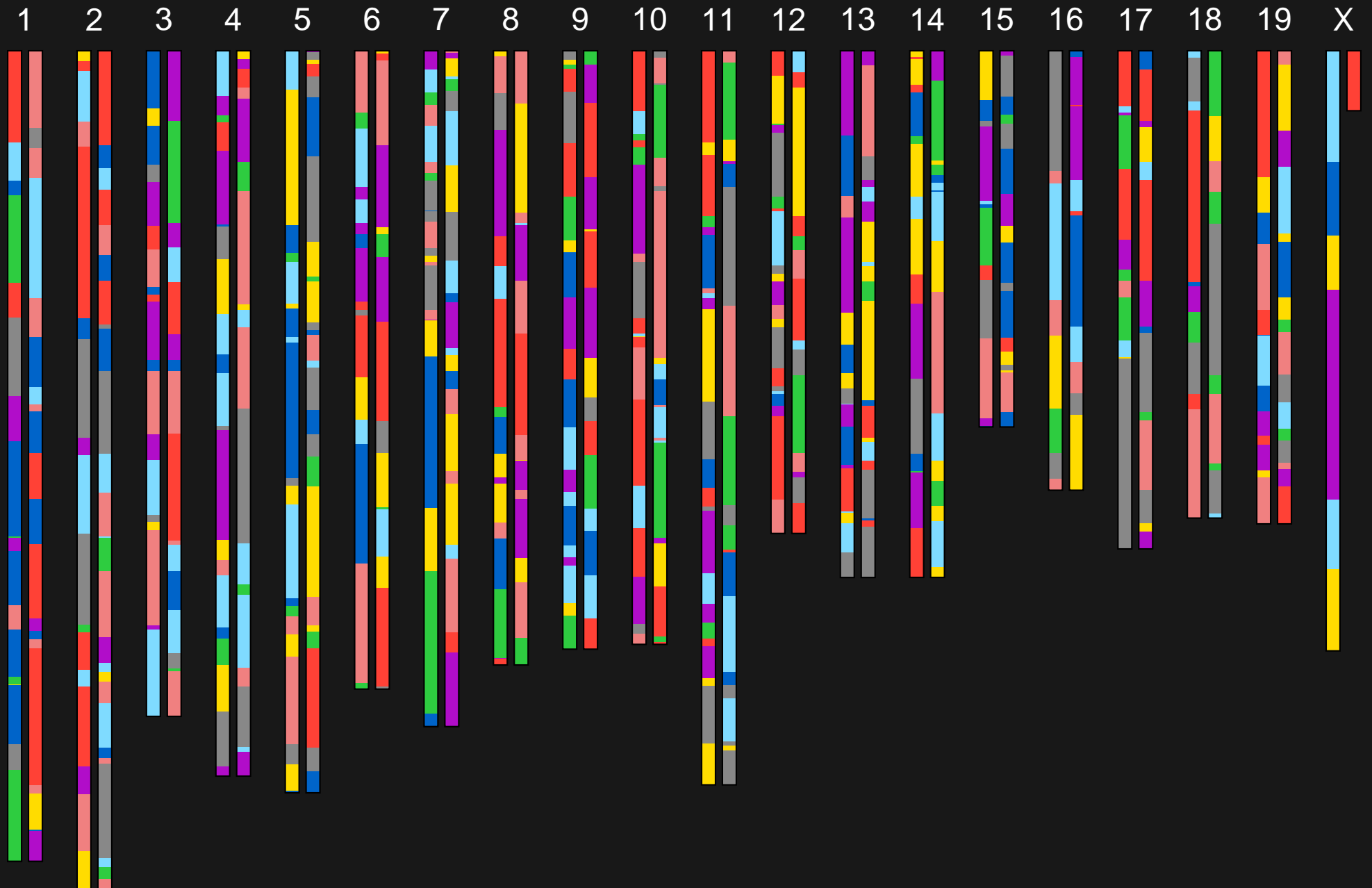
$$\begin{aligned} y_i &= \mu + \sum_k \beta_k q_{ik} + \epsilon_i & \epsilon_i &\sim \mathbf{N}(0, \sigma_e^2) \\ &= \mu + \eta_i + \epsilon_i & \eta_i &\sim \mathbf{N}(0, \sigma_p^2) \end{aligned}$$

$$\mathbf{COV}(\eta_i, \eta_j) = \sigma_p^2 (2k_{ij})$$

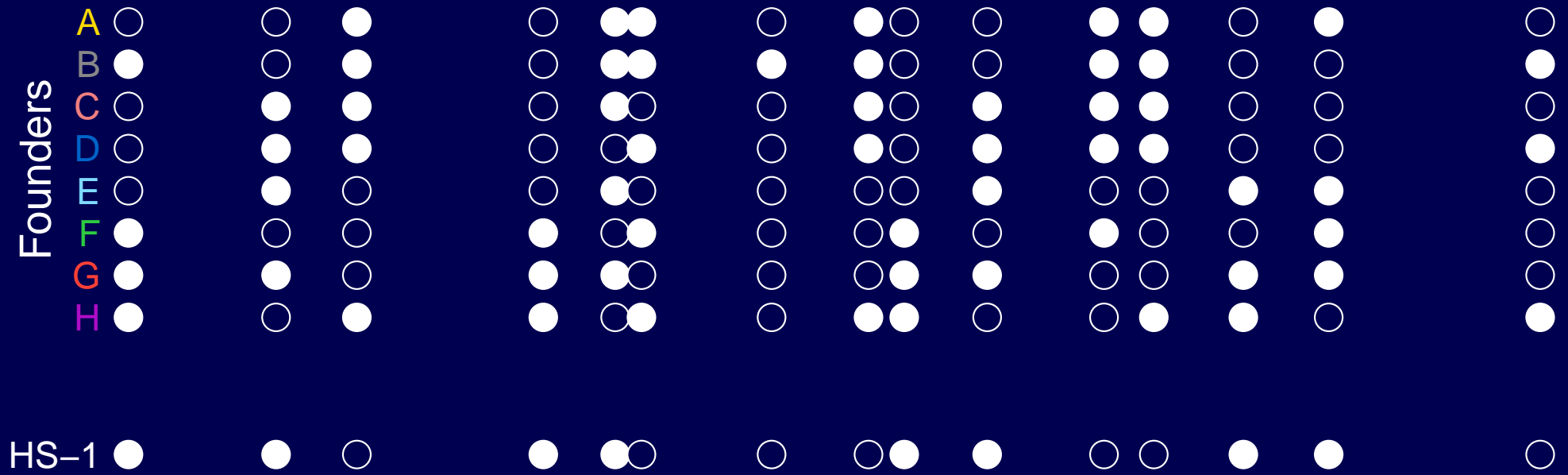
→ R

- `calc_kinship()`
- `scan1()`

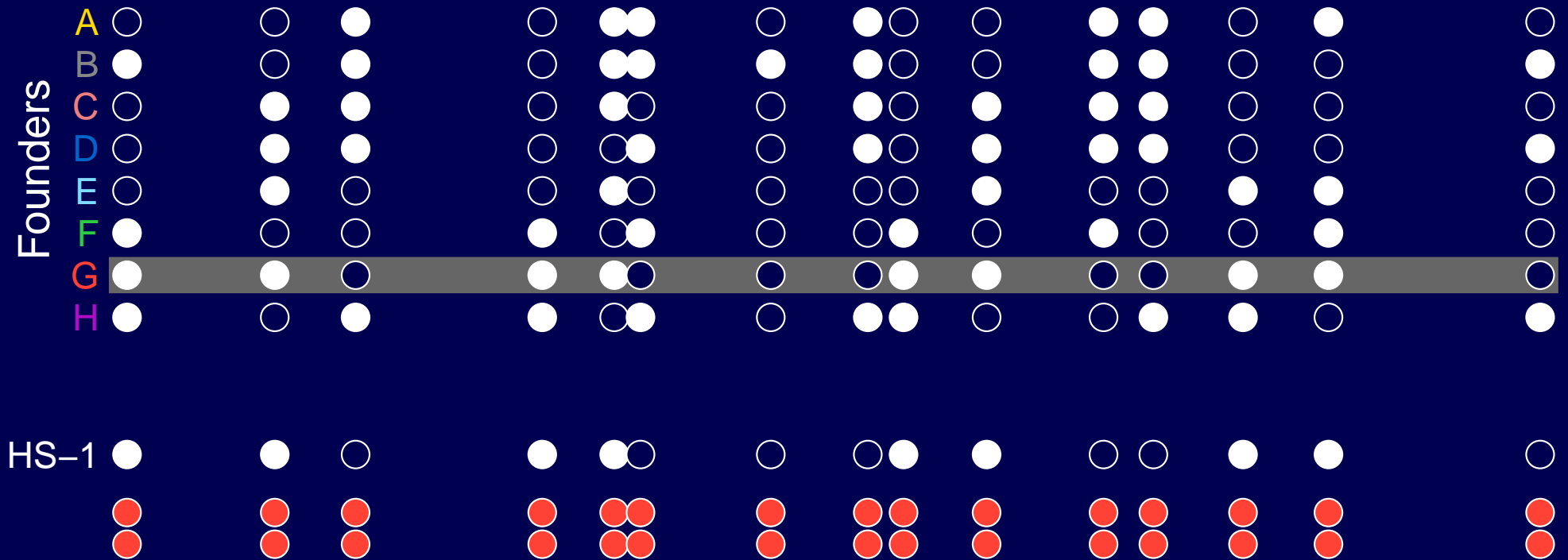
HS genome



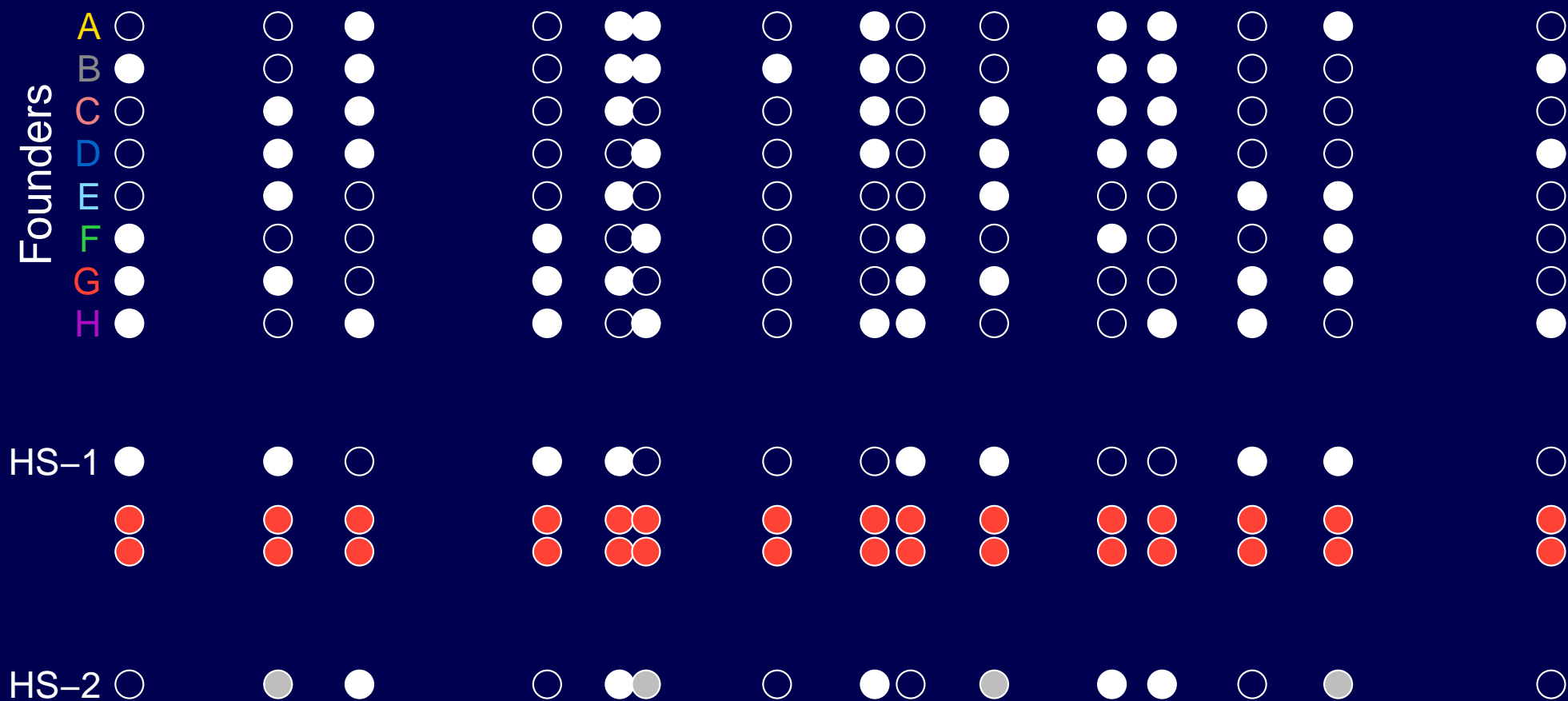
HS genotype reconstruction



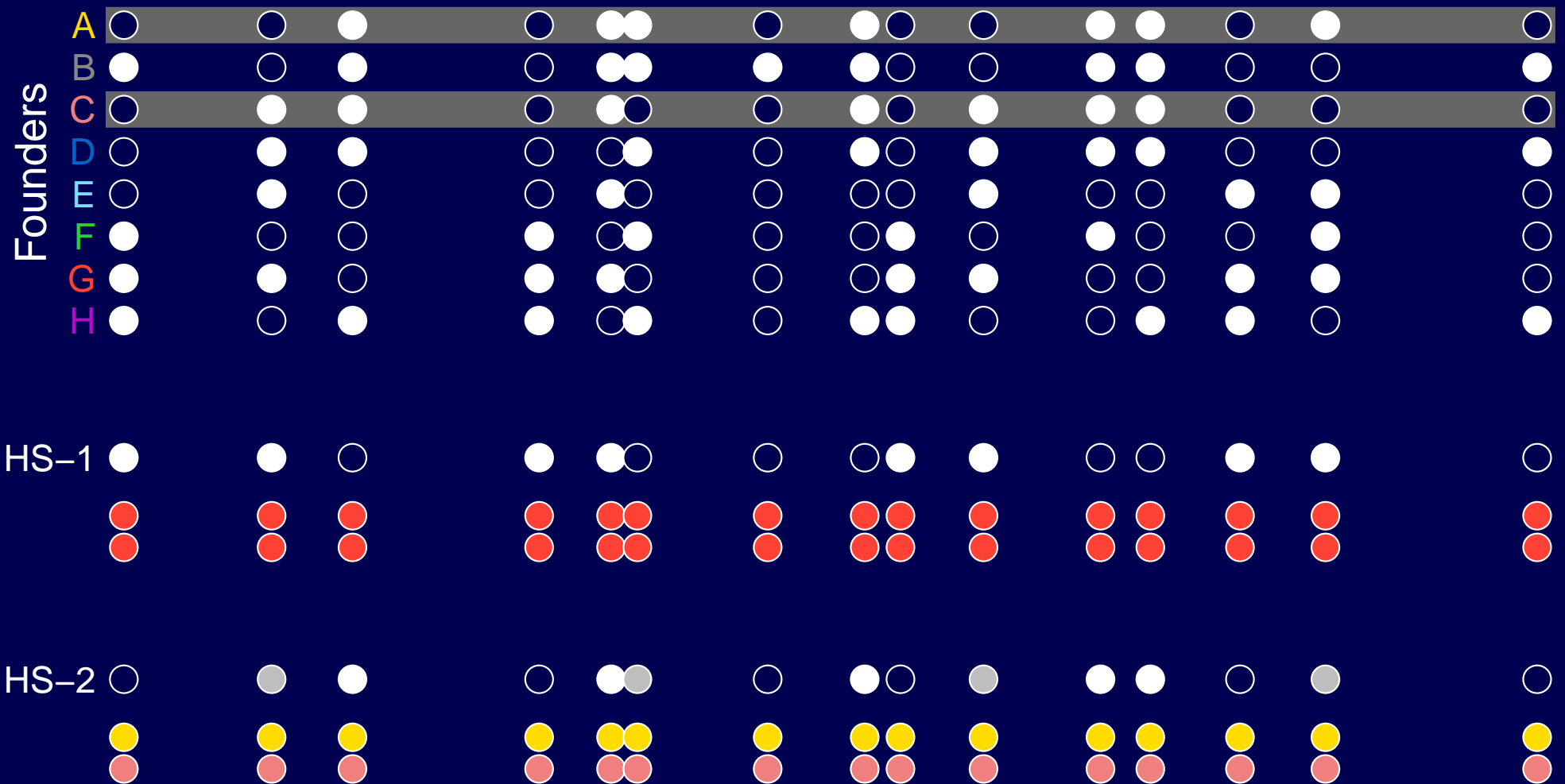
HS genotype reconstruction



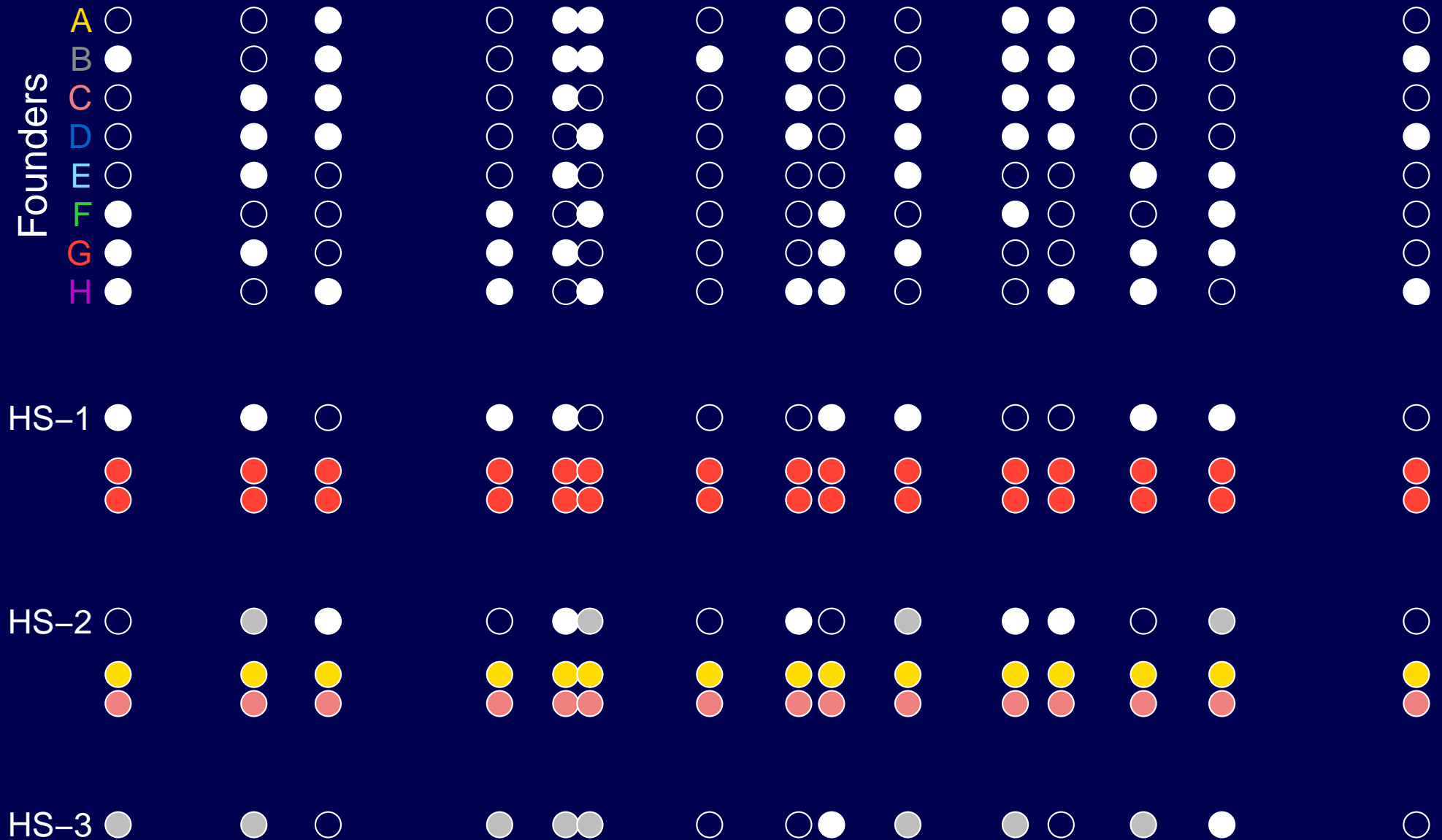
HS genotype reconstruction



HS genotype reconstruction



HS genotype reconstruction



HS genotype reconstruction

